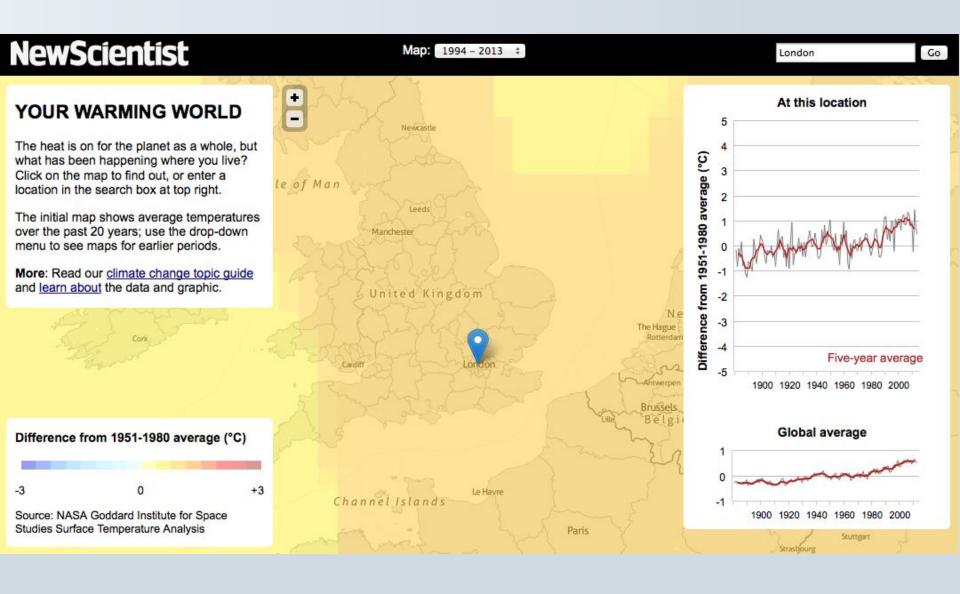
Using data in science journalism

Peter Aldhous,
Science reporter, BuzzFeed News

peter@peteraldhous.com @paldhous

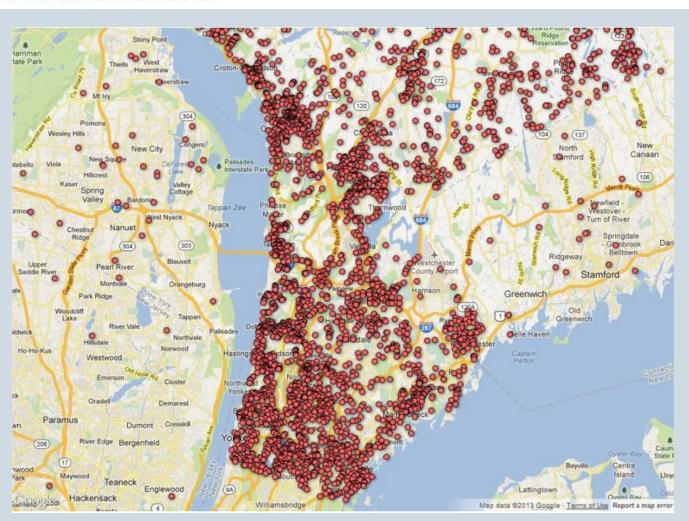
Is this data journalism?



Is this data journalism?

Map: Where are the gun permits in your neighborhood?

FILED UNDER - News / Local News | 3:51 PM, Dec. 22, 2012



Is this data journalism?

Search All NYTimes.com The New york Times U.S. Go WORLD N.Y. / REGION BUSINESS TECHNOLOGY SCIENCE HEALTH SPORTS OPINION ARTS STYLE TRAVEL **JOBS AUTOS** REAL ESTATE POLITICS EDUCATION TEXAS

BREAKDOWN Death and disarray at America's racetracks

Mangled Horses, Maimed Jockeys

The new economics of horse racing are making an always-dangerous game even more so, as lax oversight puts animal and rider at risk.

By WALT BOGDANICH, JOE DRAPE, DARA L. MILES and GRIFFIN PALMER Published: March 24, 2012

☐ 481 Comments

RUIDOSO, N.M. — At 2:11 p.m., as two ambulances waited with motors running, 10 horses burst from the starting gate at Ruidoso Downs Race Track 6,900 feet up in New Mexico's Sacramento Mountains.

Nineteen seconds later, under a brilliant blue sky, a national champion jockey named Jacky Martin lay sprawled in the furrowed dirt just past the finish line, paralyzed, his neck broken in three places. On the ground next to him, his frightened horse, leg broken and chest heaving, was minutes away from being euthanized on the track.

For finishing fourth on this early September day last year, Jacky Martin got about \$60 and possibly a lifetime tethered to a respirator.





A Jockey's Story

Jockeys are on the front lines of the risky business of horse racing. Twenty-four horses a week die at racetracks around the country.

Video by Matthew Orr/The New York Times

Data + journalism = story



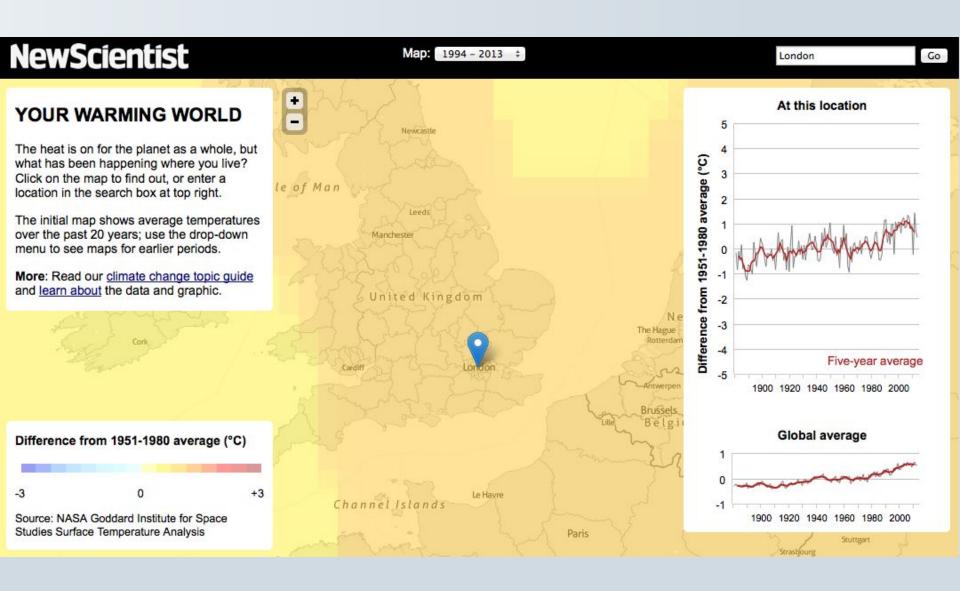
BREAKDOWN Death and disarray at America's racetracks

Mangled Horses, Maimed Jockeys

The new economics of horse racing are making an always-dangerous game even more so, as lax oversight puts animal and rider at risk.

But an investigation by The New York Times has found that industry practices continue to put animal and rider at risk. A computer analysis of data from more than 150,000 races, along with injury reports, drug test results and interviews, shows an industry still mired in a culture of drugs and lax regulation and a fatal breakdown rate that remains far worse than in most of the world.

... which can be told visually



But it's not only maps and graphs!

Data can be used in both reporting and storytelling.

And visualization can be a powerful for both finding and telling stories.

But think carefully about what you need to show to your audience. Some of the best data-driven stories contain little in the way of numbers or graphs.

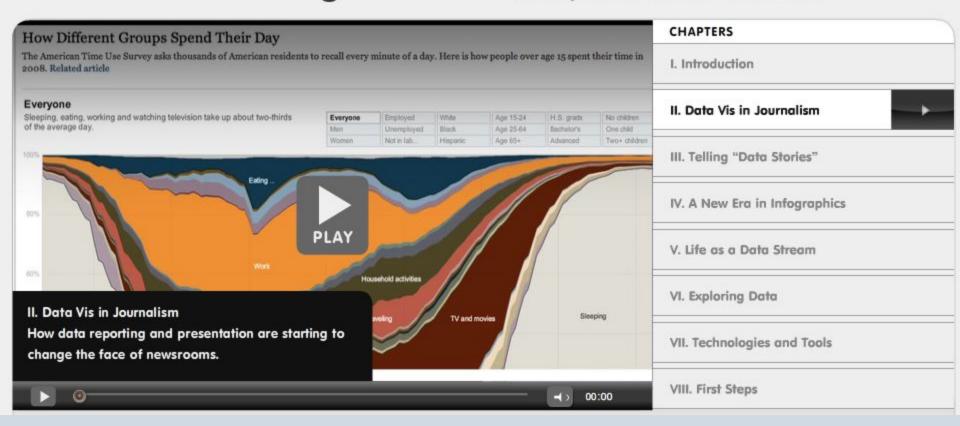
Have you ...

- Filed a request for public records?
- Worked on a story that required a spreadsheet to analyze data?
- Worked on a story that relied on querying an online database?
- Obtained a database and analyzed with database management software, or built your own database?
- Used mapping software for a story?
- Used statistical analysis for a story?
- Studied connections between people/organizations using network analysis?

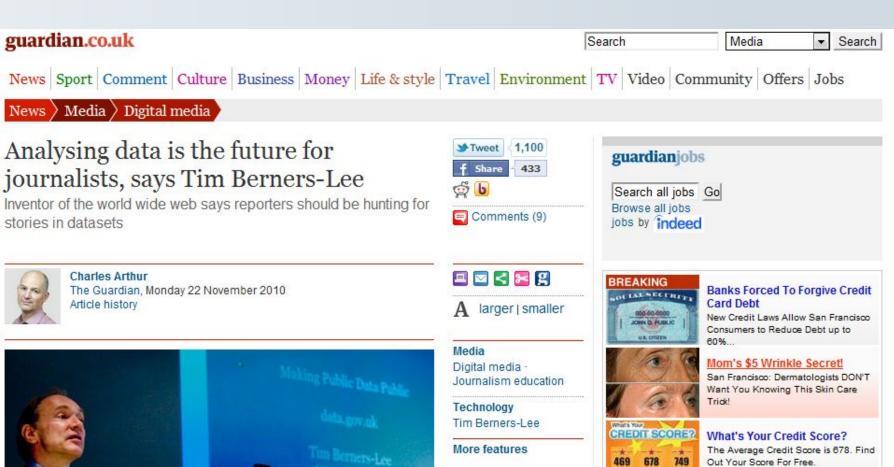
A new thing?

Journalism in the Age of Data

A video report on data visualization as a storytelling medium Produced during a 2009-2010 Knight Journalism Fellowship Total Running Time: 54 Minutes; with related information and links



A thing of the future?





Tim Berners-Lee. Photograph: Guardian

Berners-Lee: Facebook could fragment web Founder of world wide

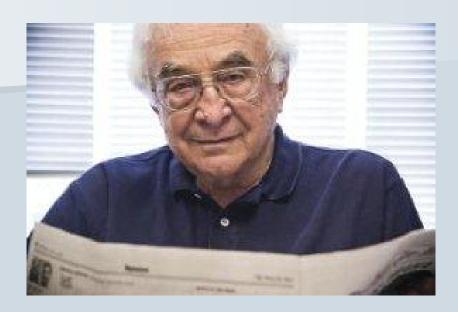
More on this story



ADS BY YABUKA.COM >

Find Out Now!

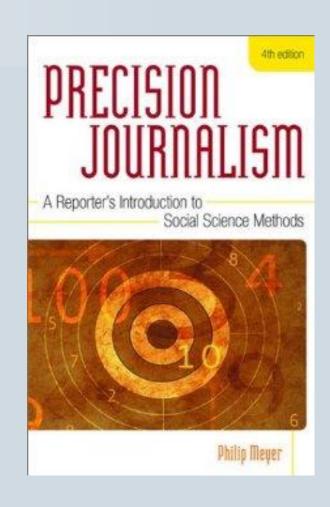
The pioneer: Philip Meyer



Now emeritus professor of journalism, University of North Carolina at Chapel Hill.

Pioneered use of quantitative methods in journalism with Knight Newspapers in 1960s.

Author of *Precision Journalism*, first published 1973.



A Pulitzer for data journalism: 1967 Detroit riot



Data: Survey conducted in the immediate aftermath of the riot.

Findings: One theory held that the rioters were reacting to being stuck at the foot of the economic ladder. Another blamed southern blacks who had moved to Detroit. But Philip Meyer showed that college graduates were as likely to have rioted as high-school dropouts, and that those born in the South were less likely to have participated.

Attention turned instead to pervasive racial discrimination in policing and housing in Detroit.

What's in it for me?

- Place your other reporting in context. Less "he said; she said."
- Visualize complex stories: Fresh understanding; new points of entry
- Find original stories, new angles

Where do I start?

Usually, with a question you want to answer, or a point you want to demonstrate.

Good data journalism rarely starts by aimlessly poking at a dataset.

Approach data like you would an interview: What do you and your audience want to know?

The data frame of mind

- When you start working on a story, think "What sources of data are available?" as well as "Who can I speak to about this?"
- Assume the data you need exists and is open to the public until proven otherwise.
- Make it a regular practice to learn about sources of data related to your beat.
- If necessary, have a plan for acquiring data at regular intervals. Some data may require public records requests.

The data frame of mind

This very different to:

"I've written my story. Now I'd better find some numbers for a graph."

A note of caution: data is often 'dirty'

Data can be seductive, but never simply assume that it is correct and consistent. Examine any data you obtain to see how it is organized, and scan for potential errors.

You will almost always need to reformat and edit data to suit your purposes; frequently you will have to do extensive data "cleaning".

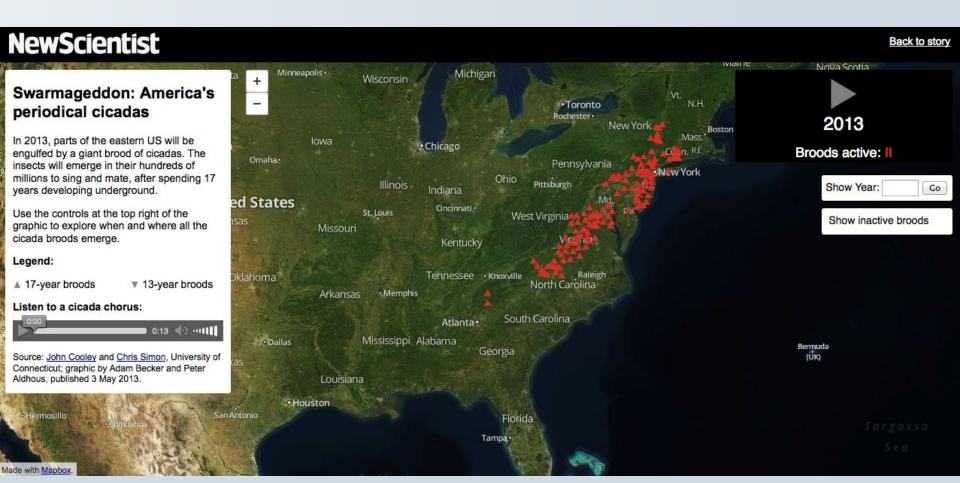
Please clean me!

d	А	В	С	D	Е	F	G	Н	1	j	K	L	М	N
	REVIEWER		11111111	MIDDLE										
1	ID	LAST NAME	FIRST NAME	INITIAL	RANK	DEGREE	SITE	STREET ADDRESS	CITY	STATE	ZIP CODE	COUNTRY	RECEIPT DATE	TYPE
2	459203	%BENN%	TERRY	L	NG	MD	RANDOLPH FAMILY PRACTICE	1918 RANDOLPH RD STE 275	CHARLOTTE	NC	28207	US	12/5/2001	DEM
3	533704	%EL-GHOROURY%	MOHAMMAD		NG	MD	NG	22201 MOROSS STE 150	DETROIT	MI	48236	US	2/11/2011	DEM
4	512096	%GUENTHER	RAINER		NG	MD	UNIVERSITATSKLINIKUM SCHLE	SCHITTENHELMSTR 12	KIEL	NG	24105	GM	11/19/2007	DEM
5	16648	%RIBOT%	THOMAS	L	NG	MD	ARNETT	2600 GREENBUSH ST	LAFAYETTE	IN	47904	US	3/7/2000	DEM
6	16648	%RIBOT%	THOMAS	L	NG	MD	ARNETT	2600 GREENBUSH ST	LAFAYETTE	IN	47904	US	5/5/2000	DEM
7	16648	%RIBOT%	THOMAS	L	NG	MD	ARNETT	2600 GREENBUSH ST	LAFAYETTE	IN	47904	US	8/21/1981	DEM
8	16648	%RIBOT%	THOMAS	L	NG	MD	ARNETT	2600 GREENBUSH ST	LAFAYETTE	IN	47904	US	9/11/2003	DEM
9	16648	%RIBOT%	THOMAS	L	NG	MD	ARNETT	2600 GREENBUSH ST	LAFAYETTE	IN	47904	US	6/9/1998	DEM
10	16648	%RIBOT%	THOMAS	L	NG	MD	ARNETT	2600 GREENBUSH ST	LAFAYETTE	IN	47904	US	5/29/1998	DEM
11	16648	%RIBOT%	THOMAS	L	NG	MD	ARNETT	2600 GREENBUSH ST	LAFAYETTE	IN	47904	US	3/12/2003	DEM
12	499673	%RICHARDSON	MARTIN	D	NG	MD	THE ROYAL MELBOURNE HOSP/	GRATTAN ST	PARKVILLE	NG	3050	AS	5/12/2006	DEM
13	534551	%TAUTH	JEFFREY		NG	MD	NG	180 MEDICAL PARK DRIVE	HOT SPRINGS	AR	71901	US	4/11/2011	DEM
14	394897	,AAVEDRA	LILLIAN	T	NG	MD	NG	1315 S ORANGE AVE STE 3E	ORLANDO	FL	32806	US	3/16/2004	DEM
15	394897	,AAVEDRA	LILLIAN	Т	NG	MD	NG	1315 S ORANGE AVE STE 3E	ORLANDO	FL	32806	US	2/5/1993	DEM
16	344230	.EVINE	KENNETH	Α	NG	MD	NG	1551 N PALM AVE	PEMBROKE PIN	IFL	33026	US	8/30/1988	DEM
17	514421	.WENS	SHEMETRA		NG	NG	MCLEAN HOSP	115 MILL STREET	BELMONT	MA	2478	US	5/15/2008	IRB
18	514421	.WENS	SHEMETRA		NG	NG	MCLEAN HOSP	115 MILL STREET	BELMONT	MA	2478	US	5/20/2008	IRB
19	514421	.WENS	SHEMETRA		NG	NG	MCLEAN HOSP	115 MILL STREET	BELMONT	MA	2478	US	1/9/2009	IRB
20	514421	.WENS	SHEMETRA		NG	NG	MCLEAN HOSP	115 MILL STREET	BELMONT	MA	2478	US	3/23/2009	IRB
21	514421	.WENS	SHEMETRA		NG	NG	MCLEAN HOSP	115 MILL STREET	BELMONT	MA	2478	US	4/27/2010	IRB
22	514421	.WENS	SHEMETRA		NG	NG	MCLEAN HOSP	115 MILL STREET	BELMONT	MA	2478	US	11/5/2009	IRB
23	514421	.WENS	SHEMETRA		NG	NG	MCLEAN HOSP	115 MILL STREET	BELMONT	MA	2478	US	3/10/2011	IRB
24	514421	.WENS	SHEMETRA		NG	NG	MCLEAN HOSP	115 MILL STREET	BELMONT	MA	2478	US	2/18/2011	IRB
25	514421	.WENS	SHEMETRA		NG	NG	MCLEAN HOSP	115 MILL STREET	BELMONT	MA	2478	US	10/16/2009	IRB
26	514421	.WENS	SHEMETRA		NG	NG	MCLEAN HOSP	115 MILL STREET	BELMONT	MA	2478	US	2/1/2010	IRB
27	514421	.WENS	SHEMETRA		NG	NG	MCLEAN HOSP	115 MILL STREET	BELMONT	MA	2478	US	3/20/2008	IRB
28	514421	.WENS	SHEMETRA		NG	NG	MCLEAN HOSP	115 MILL STREET	BELMONT	MA	2478	US	7/2/2009	IRB
29	532708	;AW	IAN		NG	MD	RIGSHOPITALET COPENHAGEN,	9 BLEGDAMSVEJ	COPENHAGEN	NG	2100	DA	11/15/2010	DEM
30	307380	??	ADAM	R	NG	MD	UNIV COLORADO/COLORADO	4200/4700 E 9TH AVE BOX C	DENVER	CO	80262	US	6/9/1999	DEM
31	307380	??	ADAM	R	NG	MD	UNIV COLORADO/COLORADO	4200/4700 E 9TH AVE BOX C	DENVER	со	80262	US	12/10/1998	DEM

But science journalists are lucky: Lots of clean, well curated data ...

ARLENE	is	number	1 of the	year	2011												
Day	Hour	Lat.	Long.	Di	r.		Spe	ed-			Wi	nd		Press	sure		Туре
28	6 UTC	19.9N	92.8W		deg		mph		kph	30	mph	45	kph	1007	mb		
28	12 UTC	20.3N	93.1W	325	deg	4	mph	7	kph	35	mph	55	kph	1006	mb		
28	18 UTC	20.7N	93.5W	315	deg	5	mph	9	kph	40	mph	65	kph	1006	mb	Tropical	Storm
29	0 UTC	21.0N	93.9W	310	deg	4	mph	7	kph	40	mph	65	kph	1005	mb	Tropical	Storm
29	6 UTC	21.2N	94.5W	290	deg	5	mph	9	kph	40	mph	65	kph	1003	mb	Tropical	Storm
29	12 UTC	21.3N	95.3W	280	deg	8	mph	12	kph	50	mph	85	kph	1000	mb	Tropical	Storm
29	18 UTC	21.4N	95.6W	290	deg	2	mph	3	kph	60	mph	95	kph	998	mb	Tropical	Storm
30	0 UTC	21.6N	96.1W	295	deg	5	mph	9	kph	60	mph	95	kph	996	mb	Tropical	Storm
30	6 UTC	21.6N	97.0W	270	deg	9	mph	14	kph	65	mph	100	kph	994	mb	Tropical	Storm
30	12 UTC	21.6N	97.3W	270	deg	2	mph	3	kph	65	mph	100	kph	993	mb	Tropical	Storm
30	18 UTC	21.5N	98.1W	260	deg	8	mph	12	kph	50	mph	85	kph	998	mb	Tropical	Storm
1	0 UTC	21.1N	98.7W	235	deg	6	mph	11	kph	35	mph	55	kph	1002	mb	Tropical	Depression
BRET	is	number	2 of the	year	2011												
*****	******	******	******	****	****	*											
Day	Hour	Lat.	Long.	Di	r.		Sp∈	ed-			Wi			Press	sure		Туре
16	6 UTC	30.7N	79.7W														
16	12 UTC	30.3N	79.4W	145	deg	4	mph	7	kph	25	mph	35	kph	1014	mb		
	Day 28 28 28 29 29 29 29 30 30 30 30 1 BRET ******* Day 16	Day Hour 28 6 UTC 28 12 UTC 28 18 UTC 29 0 UTC 29 6 UTC 29 12 UTC 29 18 UTC 30 0 UTC 30 6 UTC 30 12 UTC 30 18 UTC 1 0 UTC BRET is ************************************	Day Hour Lat. 28 6 UTC 19.9N 28 12 UTC 20.3N 28 18 UTC 20.7N 29 0 UTC 21.0N 29 6 UTC 21.2N 29 12 UTC 21.3N 29 18 UTC 21.4N 30 0 UTC 21.6N 30 6 UTC 21.6N 30 12 UTC 21.6N 30 18 UTC 21.5N 1 0 UTC 21.1N BRET is number ***********************************	Day Hour Lat. Long. 28 6 UTC 19.9N 92.8W 28 12 UTC 20.3N 93.1W 28 18 UTC 20.7N 93.5W 29 0 UTC 21.0N 93.9W 29 6 UTC 21.2N 94.5W 29 12 UTC 21.3N 95.3W 29 18 UTC 21.4N 95.6W 30 0 UTC 21.6N 96.1W 30 6 UTC 21.6N 97.0W 30 12 UTC 21.6N 97.3W 30 18 UTC 21.5N 98.1W 1 0 UTC 21.1N 98.7W BRET is number 2 of the ***********************************	Day Hour Lat. Long. Di 28 6 UTC 19.9N 92.8W 28 12 UTC 20.3N 93.1W 325 28 18 UTC 20.7N 93.5W 315 29 0 UTC 21.0N 93.9W 310 29 6 UTC 21.2N 94.5W 290 29 12 UTC 21.3N 95.3W 280 29 18 UTC 21.4N 95.6W 290 30 0 UTC 21.6N 96.1W 295 30 6 UTC 21.6N 97.0W 270 30 12 UTC 21.6N 97.3W 270 30 12 UTC 21.5N 98.1W 260 1 0 UTC 21.1N 98.7W 235 BRET is number 2 of the year ************************************	Day Hour Lat. Long. Dir. 28 6 UTC 19.9N 92.8W deg 28 12 UTC 20.3N 93.1W 325 deg 28 18 UTC 20.7N 93.5W 315 deg 29 0 UTC 21.0N 93.9W 310 deg 29 6 UTC 21.2N 94.5W 290 deg 29 12 UTC 21.3N 95.3W 280 deg 29 18 UTC 21.4N 95.6W 290 deg 30 0 UTC 21.6N 96.1W 295 deg 30 6 UTC 21.6N 97.0W 270 deg 30 12 UTC 21.6N 97.3W 270 deg 30 12 UTC 21.6N 97.3W 270 deg 30 18 UTC 21.5N 98.1W 260 deg 1 0 UTC 21.1N 98.7W 235 deg BRET is number 2 of the year 2011 **********************************	Day Hour Lat. Long. Dir 28 6 UTC 19.9N 92.8W deg 28 12 UTC 20.3N 93.1W 325 deg 4 28 18 UTC 20.7N 93.5W 315 deg 5 29 0 UTC 21.0N 93.9W 310 deg 4 29 6 UTC 21.2N 94.5W 290 deg 5 29 12 UTC 21.3N 95.3W 280 deg 8 29 18 UTC 21.4N 95.6W 290 deg 2 30 0 UTC 21.6N 96.1W 295 deg 5 30 6 UTC 21.6N 96.1W 295 deg 5 30 6 UTC 21.6N 97.0W 270 deg 9 30 12 UTC 21.6N 97.3W 270 deg 2 30 18 UTC 21.6N 97.3W 270 deg 2 30 18 UTC 21.5N 98.1W 260 deg 8 1 0 UTC 21.1N 98.7W 235 deg 6 BRET is number 2 of the year 2011 ***********************************	Day Hour Lat. Long. DirSpecial Special S	Day Hour Lat. Long. DirSpeed- 28 6 UTC 19.9N 92.8W deg mph 28 12 UTC 20.3N 93.1W 325 deg 4 mph 7 28 18 UTC 20.7N 93.5W 315 deg 5 mph 9 29 0 UTC 21.0N 93.9W 310 deg 4 mph 7 29 6 UTC 21.2N 94.5W 290 deg 5 mph 9 29 12 UTC 21.3N 95.3W 280 deg 8 mph 12 29 18 UTC 21.4N 95.6W 290 deg 2 mph 3 30 0 UTC 21.6N 96.1W 295 deg 5 mph 9 30 6 UTC 21.6N 97.0W 270 deg 9 mph 14 30 12 UTC 21.6N 97.3W 270 deg 2 mph 3 30 18 UTC 21.6N 97.3W 270 deg 8 mph 12 1 0 UTC 21.1N 98.7W 235 deg 6 mph 11 BRET is number 2 of the year 2011 **********************************	Day Hour Lat. Long. DirSpeed 28 6 UTC 19.9N 92.8W deg mph kph 28 12 UTC 20.3N 93.1W 325 deg 4 mph 7 kph 28 18 UTC 20.7N 93.5W 315 deg 5 mph 9 kph 29 0 UTC 21.0N 93.9W 310 deg 4 mph 7 kph 29 6 UTC 21.2N 94.5W 290 deg 5 mph 9 kph 29 12 UTC 21.3N 95.3W 280 deg 8 mph 12 kph 29 18 UTC 21.4N 95.6W 290 deg 2 mph 3 kph 30 0 UTC 21.6N 96.1W 295 deg 5 mph 9 kph 30 6 UTC 21.6N 97.0W 270 deg 9 mph 14 kph 30 12 UTC 21.6N 97.3W 270 deg 2 mph 3 kph 30 18 UTC 21.6N 97.3W 270 deg 2 mph 3 kph 30 18 UTC 21.5N 98.1W 260 deg 8 mph 12 kph 1 0 UTC 21.1N 98.7W 235 deg 6 mph 11 kph BRET is number 2 of the year 2011 **********************************	Day Hour Lat. Long. DirSpeed 28 6 UTC 19.9N 92.8W deg mph kph 30 28 12 UTC 20.3N 93.1W 325 deg 4 mph 7 kph 35 28 18 UTC 20.7N 93.5W 315 deg 5 mph 9 kph 40 29 0 UTC 21.0N 93.9W 310 deg 4 mph 7 kph 40 29 6 UTC 21.2N 94.5W 290 deg 5 mph 9 kph 40 29 12 UTC 21.3N 95.3W 280 deg 8 mph 12 kph 50 29 18 UTC 21.4N 95.6W 290 deg 2 mph 3 kph 60 30 0 UTC 21.6N 96.1W 295 deg 5 mph 9 kph 60 30 6 UTC 21.6N 97.0W 270 deg 9 mph 14 kph 65 30 12 UTC 21.6N 97.3W 270 deg 2 mph 3 kph 65 30 18 UTC 21.5N 98.1W 260 deg 8 mph 12 kph 50 1 0 UTC 21.1N 98.7W 235 deg 6 mph 11 kph 35 BRET is number 2 of the year 2011 **********************************	Day Hour Lat. Long. DirSpeedWilliams 6 UTC 19.9N 92.8W deg mph kph 30 mph 28 12 UTC 20.3N 93.1W 325 deg 4 mph 7 kph 35 mph 28 18 UTC 20.7N 93.5W 315 deg 5 mph 9 kph 40 mph 29 0 UTC 21.0N 93.9W 310 deg 4 mph 7 kph 40 mph 29 6 UTC 21.2N 94.5W 290 deg 5 mph 9 kph 40 mph 29 12 UTC 21.3N 95.3W 280 deg 8 mph 12 kph 50 mph 29 18 UTC 21.4N 95.6W 290 deg 2 mph 3 kph 60 mph 30 0 UTC 21.6N 96.1W 295 deg 5 mph 9 kph 60 mph 30 6 UTC 21.6N 97.0W 270 deg 9 mph 14 kph 65 mph 30 12 UTC 21.6N 97.3W 270 deg 2 mph 3 kph 65 mph 30 12 UTC 21.6N 97.3W 270 deg 2 mph 3 kph 65 mph 30 18 UTC 21.5N 98.1W 260 deg 8 mph 12 kph 50 mph 1 0 UTC 21.1N 98.7W 235 deg 6 mph 11 kph 35 mph	Day Hour Lat. Long. DirSpeed Wind 28 6 UTC 19.9N 92.8W deg mph kph 30 mph 45 28 12 UTC 20.3N 93.1W 325 deg 4 mph 7 kph 35 mph 55 28 18 UTC 20.7N 93.5W 315 deg 5 mph 9 kph 40 mph 65 29 0 UTC 21.0N 93.9W 310 deg 4 mph 7 kph 40 mph 65 29 6 UTC 21.2N 94.5W 290 deg 5 mph 9 kph 40 mph 65 29 12 UTC 21.3N 95.3W 280 deg 8 mph 12 kph 50 mph 85 29 18 UTC 21.4N 95.6W 290 deg 2 mph 3 kph 60 mph 95 30 0 UTC 21.6N 96.1W 295 deg 5 mph 9 kph 60 mph 95 30 6 UTC 21.6N 97.0W 270 deg 9 mph 14 kph 65 mph 100 30 12 UTC 21.6N 97.3W 270 deg 2 mph 3 kph 65 mph 100 30 18 UTC 21.5N 98.1W 260 deg 8 mph 12 kph 50 mph 85 1 0 UTC 21.1N 98.7W 235 deg 6 mph 11 kph 35 mph 55 BRET is number 2 of the year 2011 **********************************	Day Hour Lat. Long. DirSpeedWind 28 6 UTC 19.9N 92.8W deg mph kph 30 mph 45 kph 28 12 UTC 20.3N 93.1W 325 deg 4 mph 7 kph 35 mph 55 kph 28 18 UTC 20.7N 93.5W 315 deg 5 mph 9 kph 40 mph 65 kph 29 0 UTC 21.0N 93.9W 310 deg 4 mph 7 kph 40 mph 65 kph 29 6 UTC 21.2N 94.5W 290 deg 5 mph 9 kph 40 mph 65 kph 29 12 UTC 21.3N 95.3W 280 deg 8 mph 12 kph 50 mph 85 kph 29 18 UTC 21.4N 95.6W 290 deg 2 mph 3 kph 60 mph 95 kph 30 0 UTC 21.6N 96.1W 295 deg 5 mph 9 kph 60 mph 95 kph 30 6 UTC 21.6N 97.0W 270 deg 9 mph 14 kph 65 mph 100 kph 30 12 UTC 21.6N 97.3W 270 deg 2 mph 3 kph 65 mph 100 kph 30 18 UTC 21.5N 98.1W 260 deg 8 mph 12 kph 50 mph 85 kph 1 0 UTC 21.1N 98.7W 235 deg 6 mph 11 kph 35 mph 55 kph BRET is number 2 of the year 2011 **********************************	Day Hour Lat. Long. DirSpeed Press 28 6 UTC 19.9N 92.8W deg mph kph 30 mph 45 kph 1007 28 12 UTC 20.3N 93.1W 325 deg 4 mph 7 kph 35 mph 55 kph 1006 28 18 UTC 20.7N 93.5W 315 deg 5 mph 9 kph 40 mph 65 kph 1006 29 0 UTC 21.0N 93.9W 310 deg 4 mph 7 kph 40 mph 65 kph 1005 29 6 UTC 21.2N 94.5W 290 deg 5 mph 9 kph 40 mph 65 kph 1003 29 12 UTC 21.3N 95.3W 280 deg 8 mph 12 kph 50 mph 85 kph 1000 29 18 UTC 21.4N 95.6W 290 deg 2 mph 3 kph 60 mph 95 kph 998 30 0 UTC 21.6N 96.1W 295 deg 5 mph 9 kph 60 mph 95 kph 998 30 6 UTC 21.6N 97.0W 270 deg 9 mph 14 kph 65 mph 100 kph 994 30 12 UTC 21.6N 97.3W 270 deg 2 mph 3 kph 65 mph 100 kph 994 30 12 UTC 21.6N 97.3W 270 deg 2 mph 3 kph 65 mph 100 kph 994 30 12 UTC 21.5N 98.1W 260 deg 8 mph 12 kph 50 mph 85 kph 998 1 0 UTC 21.1N 98.7W 235 deg 6 mph 11 kph 35 mph 55 kph 1002 BRET is number 2 of the year 2011 ***********************************	Day Hour Lat. Long. DirSpeed	Day Hour Lat. Long. DirSpeedWind Pressure

...which scientists may be willing to share



The basics (OK, I have some data. What now?)

Sort

Largest to smallest; Alphabetical etc

Aggregate

Count, Sum, Mean, Median, Maximum, Minimum etc

Filter

Select a defined subset of the data

Join

Merge entries from two or more datasets based on common field(s), e.g. unique ID number, last name and first name

Think of those operations as 'interviewing' the data

Tools and stories: online databases



20 | NewScientist | 14 October 2006 www.newscientist.com www.newscientist.com 14 October 2006 | NewScientist | 27

above – the anthrax project – generated serious objections from

some of our experts.

vaccinated (13 January 2001, p 4). The scientists, hoping to control plagues of mice, engineered a mousepox virus

Tools and stories: work with databases on your own computer

Revealed: Pfizer's payments to censured doctors

) 16:18 22 April 2010 by Peter Aldhous and Jim Giles

They are billed as "healthcare professionals who spend years building expertise in their fields". Using materials firmly grounded in science, they educate their peers in the risks and benefits of drugs.

This is how Pfizer, the pharmaceuticals giant, describes the experts it hires to lead educational forums in which doctors are lectured on the use of its products.

Yet New Scientist has found that some of Pfizer's experts have been disciplined for deficiencies in patient care, while others have been reprimanded for how they conducted drug research trials.

The findings add to a growing controversy surrounding the pharmaceutical industry's efforts to market drugs by influencing patterns of prescribing.

Unknown influence

Doctors paid to educate peers are a particular worry, argues Sidney Wolfe of consumer advocacy group Public Citizen in Washington DC. "They are doing things that may be influencing your doctor and you have no way of knowing about it," he says. "It's made worse by the fact that some of them have been disciplined."

Data: Pfizer's records of payments to doctors, scraped from the web. Data on disciplinary actions from state medical boards in four largest states; FDA warning letters to clinical investigators

Findings: Some of Pfizer's "experts" had questionable records for patient safety



NewScientist

Tools and stories: databases











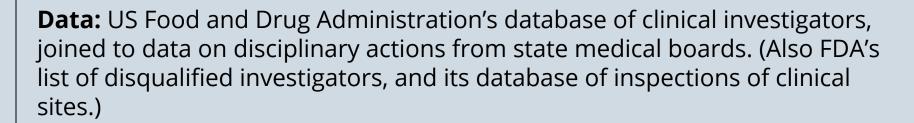
18 min



Why Are Dope-Addicted, Disgraced Doctors Running Our Drug Trials?

By Peter Aldhous

Photographs by Grant Cornett



Findings: Dozens of doctors selected to work on clinical trials over the past five years had previously been censured by state medical boards for problems with patient care; some had their own problems with substance abuse.

Tools and stories: databases

Home | Opinion | Health | Science in Society | News

My 'non-human' DNA: a cautionary tale

-) 15:02 26 August 2009 by Peter Aldhous
-) For similar stories, visit the Genetics Topic Guide

"This is a strange question, but are you sure this is Homo sapiens?"

It's not every day that an expert queries whether your DNA is human, so when I received this comment by email earlier this month I was somewhat bemused.

I am not in fact the result of a coupling between human and alien, nor the product of some twisted genetic experiment. Instead, Blaine Bettinger, who blogs as The Genetic Genealogist, had been baffled by a DNA profile generated in error by deCODEme, a leading commercial "personal genomics" service provided by Decode Genetics in Reykjavik, Iceland. The false profile seems to be the fault of a software bug.

No harm was done, but the incident serves as a cautionary tale for personalised medicine. As we move towards a future in which readouts from our genomes will routinely be queried by computer systems to help doctors make important clinical decisions, similar glitches could cause prescribing errors – with patients being given drugs at the wrong dose, drugs that won't work, or ones that could even trigger serious side effects in people with a

Data: Downloads of my own genetic scans, performed by 23andMe and DeCode Genetics. Corresponding data for my DNA markers read from the same companies' online "genome browsers".

Findings: DeCode had a glitch in its database software that could cause the presentation of an erroneous mitochondrial DNA profile in its genome browser.



NewScientist



NewScientist

	A B	С	D	Е	F	G	Н		J	K	L	M	N
1	Position (Human NCBI Build Record 36)	Position (CRS)	23andMe ID	23andMe	23andMe Download Genotype	23andMe Browser	23andMe Consistency	DeCodeMe ID		DeCodeMe	DeCodeMe Browser Genotype	DecodeMe Consistency	Consistency between 23andMe and DeCodeMe Downloads
20	19 2887	2885	rs2854130	C/T	T	T	Consistent	MitoT2887C	C/T	T	No data	Data in down	Consistent
21	20 3012	3010	rs3928306	A/G	A	Α	Consistent	MitoG3012A	A/G	A	Α	Consistent	Consistent
22	21 3198	3197	rs2854131	C/T	T	T	Consistent	MitoT3198C	C/T	T	T	Consistent	Consistent
23	22 3349	3348	rs41423746	A/G	A	A	Consistent	MitoA3349G	A/G	A	A	Consistent	Consistent
24	23 3395	3394	rs41460449	C/T	T	T	Consistent	MitoT3395C	C/T	T	C	Mismatch	Consistent
25	24 3481	3480	rs28358584	A/G	Α	Α	Consistent	MitoA3481G	A/G	A	G	Mismatch	Consistent
26	25 3595	3594	rs2854134	C/T	С	С	Consistent	MitoC3595T	C/T	C	T	Mismatch	Consistent
27	26 3667	3666	rs28357968	A/G	G	G	Consistent	MitoG3667A	A/G	G	G	Consistent	Consistent
28	27 3721	3720	rs41355750	A/G	Α	A	Consistent	MitoA3721G	A/G	A	G	Mismatch	Consistent
29	28 3916	3915	rs41524046	A/G	G	G	Consistent	MitoG3916A	A/G	G	G	Consistent	Consistent
30	29 3919	3918	rs28357972	A/G	G	G	Consistent	MitoG3919A	A/G	G	G	Consistent	Consistent
31	30 3971	3970	rs28357973	C/G/T	С	С	Consistent	MitoC3971T	C/T	С	T	Mismatch	Consistent
32	31 3993	3992	rs41402945	A/T	С	С	Consistent	MitoC3993T	C/T	С	T	Mismatch	Consistent
33	32 4025	4024	i1000001	A/G	Α	Α	Consistent	MitoA4025G	A/G	A	Α	Consistent	Consistent
34	33 4337	4336	i3001462	C/T	T	T	Consistent	MitoT4337C	C/T	T	С	Mismatch	Consistent
35	34 4562	4561	i1000011	C/T	T	T	Consistent	MitoT4562C	C/T	T	С	Mismatch	Consistent
36	35 4770	4769	rs3021086	A/G	G	G	Consistent	MitoG4770A	A/G	G	Α	Mismatch	Consistent
37	36 4821	4820	rs28357977	A/G	G	G	Consistent	MitoG4821A	A/G	G	G	Consistent	Consistent
38	37 4825	4824	rs28357978	A/G	A	Α	Consistent	MitoA4825G	A/G	A	No data	Data in down	Consistent
39	38 4884	4883	rs28357979	C/T	С	С	Consistent	MitoC4884T	C/T	С	Т	Mismatch	Consistent
40	39 4918	4917	rs28357980	A/G	A	Α	Consistent	MitoA4918G	A/G	Α	Α	Consistent	Consistent
41	40 4978	4977	rs28357981	C/T	T	T	Consistent	MitoT4978C	C/T	Т	С	Mismatch	Consistent

Tools and stories: putting data onto maps

Donald Trump Vs. Sea Level Rise

The Republican presidential frontrunner dismisses climate change as a hoax. Here's what rising sea levels could do to the buildings that bear his name.



BuzzFeedNews

Tools and stories: putting data onto maps

The Seattle Times

Logging and landslides: What went wrong?

When Weyerhaeuser began clear-cutting the Douglas firs on the slopes surrounding Little Mill Creek, local water officials were on edge. Some of these lands had slid decades ago, after an earlier round of logging. They worried new slides could dump sediments into the mountain stream and overwhelm a treatment plant. Those fears came true last December.

By Hal Bernton and Justin Mayo

Seattle Times staff reporters

BOISTFORT VALLEY, Lewis County — When Weyerhaeuser began clear-cutting the Douglas firs on the slopes surrounding Little Mill Creek, local water officials were on edge.

Some of these lands had slid decades ago, after an earlier round of logging. They worried new slides could dump sediments into the mountain stream and overwhelm a treatment plant.

Those fears came true last December when a monster storm barreled in from the Pacific, drenching the mountains around the Chehalis River basin and touching off hundreds of landslides. Little Mill Creek, filled with mud and debris, turned dark like chocolate syrup.

More than three months passed before nearly 3,000 valley residents could drink from their taps again.

"I have never seen anything like this before, and I hope I never do again," said Fred Hamilton, who works for the Boistfort Valley Water Corp.

State forestry rules empower the Department of Natural Resources (DNR) to restrict logging on



Data: GIS data on clear-cuts, landslides and prior studies of the hazards from the state Dept of Natural Resources; logging company Weyerhaeuser's logging permits.

Findings: With little scrutiny from state geologists, Weyerhaeuser was allowed to clear-cut unstable slopes.

Using mapping software, the reporters showed that clear-cut sites that had at least half of their acreage in a moderate- to high-hazard zone accounted for a disproportionate number of landslides in December 2007 storms.

Statistical analysis, network analysis, etc:

Not just for our scientist sources



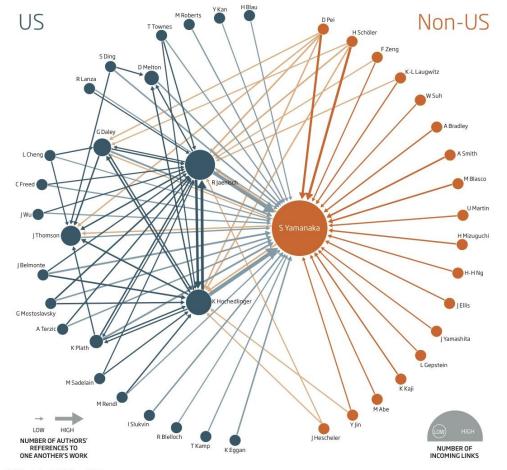
NewScientist

SPECIAL REPORT

THE STEM CELL WARS

When a Nobel prize is up for grabs, do scientists across the globe compete on a level playing field? **Peter Aldhous** investigates

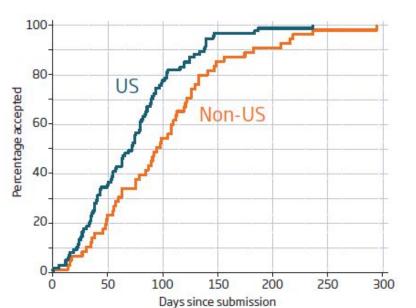
The most influential players in cellular reprogramming are revealed by recording how many times the scientists have referred to each other's work. Each link shows where one researcher cited another four or more times in papers in leading journals (for analysis, see "The strongest link", below right)



12 | NewScientist | 12 June 2010

100612_N_Special_Report.indd 12 8/6/10 14:10:48

What's the hold-up? In a sample of 148 papers from high-profile journals, those from scientists outside the US took longer to be accepted for publication



Data: Time-to-acceptance for papers involving "induced pluripotent stem cells" – an exciting alternative to embryonic stem cells which later won their discoverer a Nobel prize.

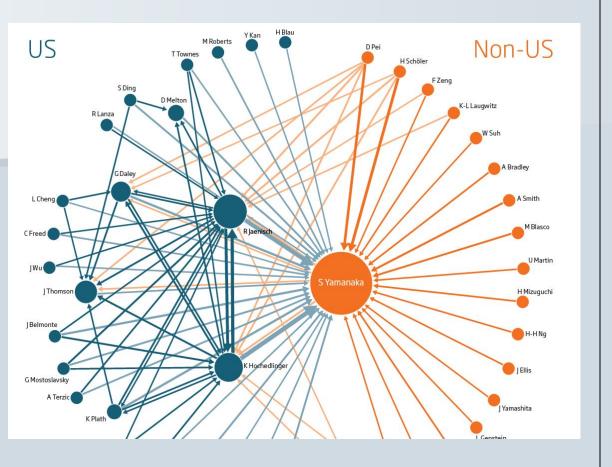
Data preparation and analysis:

Searches and downloads from Web of Science database. Annotation of spreadsheet to document time of receipt, acceptance and publication, and location of primary author. Statistical and graphical analysis of time-to-acceptance data. See methods.

Findings: Papers from corresponding authors outside the US took significantly longer to be accepted for publication



NewScientist





Data: Citations between primary authors of papers on iPS cells.

Data preparation and analysis: Citation information extracted from Web of Science using academic bibliometric software; database queries to link the citations by primary author. Manual checking for errors. Then analysis of network graph. See methods.

Findings: The citation network graph maps influence and connections in the field. Does this help explain why non-US scientists seem to be losing the race to publish?



Data: Metadata for 34,000+ papers published in PNAS from 2004-2013, plus citation counts, scraped from the journal's website.

Findings: Few academy members "contribute" papers at close to the maximum rate, but this group includes several members of the journal's editorial board. Contributed papers are cited less often than those reviewed in the normal way - although the gap has narrowed in recent years.



The inside track

Members of the US National Academy of Sciences have long enjoyed a privileged path to publication in the body's prominent house journal. Meet the scientists who use it most heavily.

BY PETER ALDHOUS

n April, the US National Academy of Sciences elected 105 new members to its ranks. Academy membership is one the most prestigious honours for a scientist, and it comes with a tangible perk: members can submit up to four papers per year to the body's high-profile journal, the venerable Proceedings of the National Academy of Sciences (PNAS), through the 'contributed' publication track. This unusual process allows authors to choose who will review their paper and how to respond to those reviewers' com-

For many academy members, this privileged path is central to the appeal of PNAS. But to some scientists, it gives the journal the appearance of an old boys' club. "Sound anachronistic? It is," wrote biochemist Steve Caplan of the University of Nebraska, Omaha, in a 2011 blog post that suggested the contributed track could be used as a "dumping ground" for certain

papers. Editors at PNAS have strived to dispel that perception.

With PNAS currently celebrating its centenary, the news team at Nature decided to examine the contributed track, both to assess its scientific impact and to see which members use it most heavily and why. After analysing a decade's worth of PNAS papers, we found that only a small number of scientists have used the track at close to the maximum allowable rate. The group includes some of the biggest names in science, and six of them sit on the editorial board at the journal. These scientists say the main motivator for using the contributed track is an intense frustration with the peer-review process at other high-profile journals, which they argue has become excessive and labori-

Our analysis also suggests that the efforts by PNAS to prevent abuse of the contributed track and to boost the quality of papers published by this route are bearing fruit. Although contributed PNAS papers attract fewer citations than those handled through the journal's standard review process, the gap has narrowed in recent years. "We have worked really hard at this," says Alan Fersht, a biophysicist at the University of Cambridge, UK, one of PNAS's associate editors and a heavy user of the contributed track.

A PRIVILEGE TO PUBLISH

An inside track to publication for academy members rests deep in PNAS's DNA. The journal was established in 1914 with the explicit goal of publishing members' "more important contributions to research" in addition to "work that appears to a member to be of particular importance". That remit led to the creation of two publishing tracks: contributed and 'communicated' papers -- manuscripts sent by non-members to colleagues in the academy, who shepherded them through review.

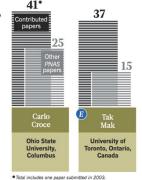
Who are the power users?

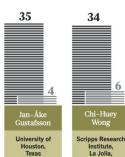
Just 13 members of the US National Academy of Sciences consistently published three or more papers per year in the 'contributed track' at PNAS during the past decade. 'Other' papers include direct submissions, reviewed in the normal way, and papers contributed or communicated by other

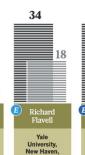


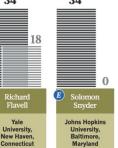
Nobel prizewinner

Member of PNAS Former member of editorial board









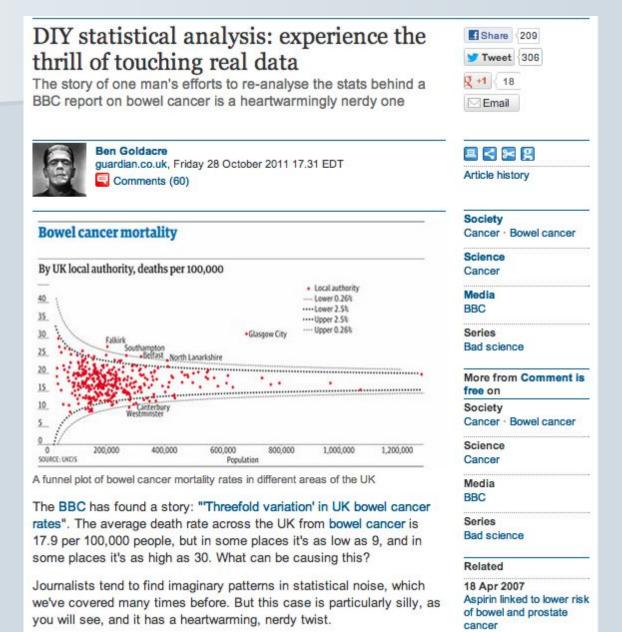


Data preparation and analysis: After the web scraping, extensive data cleaning to remove variants of authors' names, giving one name format for each academy member. Database queries to count papers of different types from each academy member. Statistical analysis to analyse citation rates of different classes of paper. See methods.

TABLE 1

		Contributed papers			Communicated papers					
Period	Effect size	95% confidence interval	p	Effect size	95% confidence interval	P				
2004-11	-4.43%	-5.39% to -3.47%	<0.001	-0.53%	-1.70% to 0.65%	0.37				
2004	-6.60%	-9.39% to -3.72%	<0.001	-3.45%	-6.37% to -0.44%	<0.05				
2005	-7.39%	-10.01% to -4.70%	<0.001	-1.02%	-3.94% to 1.99%	0.5				
2006	-7.01%	-9.63% to -4.32%	<0.001	-1.42%	-4.23% to 1.47%	0.33				
2007	-5.81%	-8.52% to -3.01%	<0.001	-3.67%	-6.48% to -0.78%	<0.05				
2008	-3.41%	-6.14% to -0.60%	<0.05	-0.12%	-3.25% to 3.11%	0.94				
2009	-1.89%	-4.65% to 0.94%	0.19	1.53%	-1.85% to 5.03%	0.38				
2010	-3.48%	-6.06% to -0.83%	<0.05	-2.50%	-6.58% to 1.76%	0.25				
2011	-1.14%	-3.70% to 1.48%	0.39	N/A	N/A	N/A				

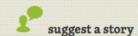
Beware running with scissors: Seek expert help if you need rigorous statistical analysis!

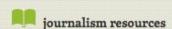


Even Nate Silver can get his fingers burned ...

Tracker

Peer review within science journalism









Nate Silver's new FiveThirtyEight dishes out statistical nonsense on health coverage.



Author: Paul Raeburn



Share



Nate Silver's fivethirtyeight.com relaunched yesterday at its new home--ESPN--vowing to focus its coverage on five areas: politics, economics, life, sports--and science.

The inclusion of science was a surprise to me. And possibly a mistake, unless FiveThirtyEight can quickly improve the quality of the "science" it's publishing. The lead story on the relaunched site's first day--"Finally, a Formula for Decoding Health News"--was abysmal.

FOLLOW US

Follow @KSJTracker on Twitter!

CATEGORIES

View All

About Journalism

Environment & Energy Stories

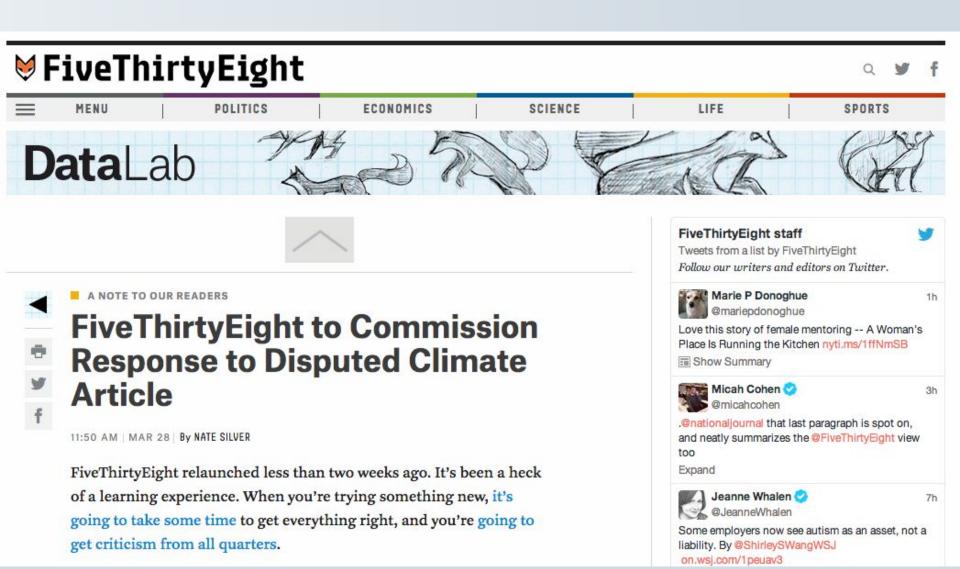
German Language Media

Health & Medicine Stories

Rastreador Científico en

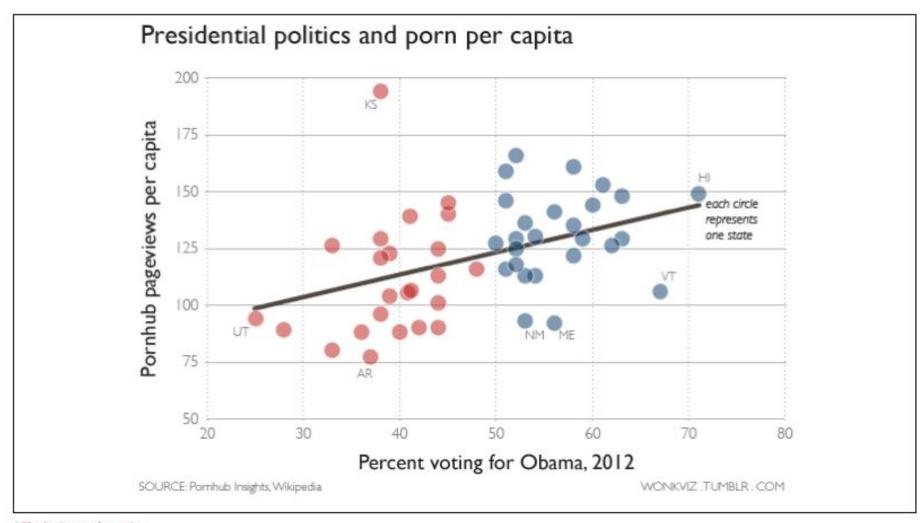
Español

... and be forced to backtrack



DISTRUST YOUR DATA

Jacob Harris on Six Ways to Make Mistakes with Data



(Chris Ingraham)

Read this before committing an act of data journalism!

Why aren't we seeing more data-driven science journalism?

Some basic data resources for science reporters

Scientific literature

PubMed

http://www-ncbi-nlm-nih-gov.oca.ucsc.edu/pubmed

Web of Science

http://bit.ly/YmJtGM

Google Scholar

http://scholar.google.com/
http://scholar.google.com.oca.ucsc.edu/

Agricola

http://agricola.nal.usda.gov/
http://agricola.nal.usda.gov.oca.ucsc.edu/

PsycINFO

http://www.apa.org/pubs/databases/psycinfo/index.aspx http://search.proquest.com.oca.ucsc.edu/psycinfo/advanced

Patents

European Patent Office

http://worldwide.espacenet.com/advancedSearch?locale=en_EP
Search European patent applications, WIPO patent applications, or all patents across 90+ nations

US Patent and Trademark Office

http://patft.uspto.gov/

Search issued patents and published applications

For more information about a patent (diagrams, history of correspondence with patent office etc), enter patent or application number here:

http://portal.uspto.gov/external/portal/pair

Google Advanced Patent Search

<u>http://www.google.com/advanced_patent_search</u>
Clear presentation of claims, abstract, diagrams etc.

Grant funding

National Institutes of Health RePORTER

http://projectreporter.nih.gov/reporter.cfm
Grants from NIH, with links to papers and patents arising from projects

National Science Foundation

http://www.nsf.gov/awardsearch/advancedSearch.jsp

Research.gov

http://www.research.gov/research-portal/appmanager/base/desktop?_nfpb=true&_ eventName=viewQuickSearchFormEvent_so_rsr

Also includes NASA grants, And note option to download **all** grants by year, from 2007 onwards

Clinical trials

ClinicalTrials.gov

http://www.clinicaltrials.gov

Information on more than 239,000 clinical trials in the US and beyond

International Standard Randomised Controlled Trial Number Register

http://www.isrctn.com

Information on more than 15,000 registered clinical trials

WHO International Clinical Trials Registry Platform

http://apps.who.int/trialsearch/default.aspx

Incorporates data from ClinicalTrials.gov, ISRCTN, and national trials registries

Using data in science journalism

Peter Aldhous,
Science reporter, BuzzFeed News

peter@peteraldhous.com @paldhous